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BY

GENERALIZED SEQUENTIAL

UNCONSTRAINED METHODS

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Contents

1. INTRODUCTION[1]

Our objective is to provide a more general theoretical basis for those methods for solving constrained minimization problems that are based on successive unconstrained minimizations of a parametric auxiliary function. In order to do this, we first give basic defining properties of a general auxiliary function and obtain a proof of local convergence for the mildly regulated nonconvex problem. We proceed from this general result to develop a family of auxiliary functions by giving the general function more and more structure, eventually being led to the general form of the " penalty" functions that conventionally have been utilized. Similarly, the problem structure is increasingly specialized and we deal finally with the convex problem. As expected, stronger results are obtained as additional structure is assumed.

1] This paper is based on material developed in [1] and essentially constitutes a generalization of results originally obtained in [5].

The problem of interest is the general mathematical programming problem

(A)   minimize $f(x)$

   subject to $g_i(x) \geq 0$, $i=1, 2, \ldots, p$ and

   $h_j(x) = 0$, $j=1, 2, \ldots, q$, where $x \in E^n$.

Since we wish to assume little more than the continuity of the problem functions for many of our results, we must account for local minima in problem (A). Even with local minima present, we could still take the point of view that only global minima are of interest, and concentrate on analyzing what is needed to bring about convergence to a global solution. However, our point of view is to attribute importance to the determination of any local minimum objective function value v^* of problem (A), and to determine at least one local minimum x^* such that $f(x^*) = v^*$. Global results follow as a byproduct of local results in this approach.

In view of this objective, we must guarantee that a sequence of points will be generated converging to any set of local minima yielding a given objective function value. For the techniques that we shall develop here, a key requirement is that, in the interior of a specified domain of interest, each such set contains a suitably prescribed closed subset. In the development given in [5], it is assumed that any such set is compact. We shall utilize this assumption, concentrating on generating bounded sequences converging to finite local solutions of (A).

We address ourselves to developing methods that generate a sequence of points converging to a local solution of (A) by means of successive unconstrained minimizations of a parametrized auxiliary function, over a specified sequence of values of the involved parameters. Also, we restrict our attention to auxiliary functions which, under

suitable conditions, converge in value to the value of the objective function $f(x)$ at a local minimum of (A). In our development, we are less concerned with the form of the penalty function and concentrate on those properties the function must possess in order to define a convergent sequential unconstrained minimization scheme. Our results are considerably more general than those obtained previously.

Particular realizations of sequential unconstrained methods in the class we shall develop have been only recently validated, by Parisot [7] for the linear problem, by Pietrzykowski [8], Fiacco and McCormick [2], [3], [4], and Pomentale [9] for the convex problem, and by Stong [10], Zangwill [12] and Fiacco and McCormick [5] for nonconvex problems as well. Many of these results, and in particular the principal convergence theorems, will be largely subsumed in the development following.

In Section 2, we define a generalized auxiliary function and prove that it yields a sequence of local unconstrained minima whose limit points are local solutions of the problem, under mild regulatory conditions. The function is defined in such a manner that, if the set of points satisfying a specified constraint has a nonempty interior, then the constraint can be enforced for all points in the minimizing sequence.

In Section 3, this function is somewhat specialized to yield the general forms of "interior-feasible" and "exterior-feasible" auxiliary functions. The convergence theorems for these functions follow as immediate corollaries of the above general result.

Further specializations lead to the functions defined and developed in [5]. These subsume the class of interior point and exterior point "penalty" functions that have conventionally been utilized in this approach. A stronger characterization of convergence is given

for these functions, in Section 4.

For the convex programming problem, the interior and exterior penalty functions can be defined to be convex, thereby leading to global and dual results. These were developed in [5] and are summarized in Section 5.

2. The Generalized Auxiliary Function for the Mixed Algorithm

For convenience, we shall write problem (A) in the form

(B)   minimize $f(x)$ subject to $g_i(x) \geq 0$, $i = 1, 2, \ldots, q$.

The following definitions and results will be used. The first is a trivial extension of the fact that a continuous function attains its minimum on a compact set.

Lemma 2.1.

If $F(x)$ is continuous in the nonempty interior of a compact set V and such that possibly $F(y) = + \infty$ for $y \in (V - V^o)$, then there exists a finite value v and a point $\overline{x} \in V$ such that $F(\overline{x}) = v = \min_V F(x)$.

Definition.

A point $x^* \in G \equiv \{x \mid g_i(x) \geq 0, \ i = 1, \ldots, q\}$ is a _local minimum_ of problem (B), if there exists an open set N such that $x^* \in N$ and $f(x^*) = \min_{G \cap N} f(x)$. If N can be selected such that $f(x^*) = \min_N f(x)$, x^* is referred to as an _unconstrained local minimum_ of (B). Otherwise, x^* is called a _constrained local_ minimum of (B).

The lemma following will be used in proving the convergence of a minimizing sequence to a prescribed compact set of minima yielding a given local minimum value of $f(x)$ in problem (B).

For significant additional generality, we require the definition of an "isolated" set.

Definition.

A nonempty set $A^* \subset A$ is called an isolated set of A if there exists a closed set E such that $E^o \supset A^*$ and such that if $x \in (E - A^*)$, then $x \notin A$.

We assume continuity of the functions of problem (B) and consider an isolated set A* of local minima A yielding a particular local minimum value, say $f(x) = v*$. The following result asserts that there exists a <u>compact</u> set S such that $S^o \supset A*$, and such that the set of <u>global</u> minima of

(B*)   minimize $f(x)$ subject to $x \in G \cap S$

is given by A*.

Let $A \equiv \{x \mid f(x) = v* \text{ and } x \text{ is a local solution of (B)}\}$.

Lemma 2.2. Existence of Compact Perturbation Set.

If the functions of (B) are continuous and if A* is a nonempty compact isolated set of A, then there exists a compact set S such that $S^o \supset A*$ and, if $\bar{x} \in G \cap S$ and $\min_{G \cap S} f(x) = f(\bar{x})$, then $\bar{x} \in A*$.

Proof. By assumption there exists a closed set E such that $E^o \supset A*$ and $x \in (E-A*)$ implies $x \notin A$. Since A* is compact, it follows that there exist compact sets S_k such that $A* \subset S_k^o \subset E$ and such that $\{S_k\} \downarrow A*$. If the conclusion of the lemma is false, then $\min_{G \cap S_k} f(x) = f(x^k) \leq v*$, with $x^k \in G \cap S_k$ and $x^k \notin A*$ for every k. Since A* is closed and since $\{x^k\}$ must contain a convergent subsequence $\{x^{k_j}\}$, it follows that $x^{k_j} \to \bar{x} \in A*$.

If $f(x^{k_j}) < v* = f(\bar{x})$ for all j, the above implies that $\bar{x}$ is not a local minimum of (B), a contradiction of the definition of A*. If $f(x^{k_j}) = v*$ for some $j = \bar{j}$ then, by construction of $\{S_k\}$, v* must be the minimum value of $f(x)$ in $G \cap S_{k_{\bar{j}}}$ for $j \geq \bar{j}$. But since $x^k \in S_{k_{\bar{j}}}^o$ for k large enough, this means x^{k_j} is a local solution of (B) with

value v* for j large enough; i.e., $x^{k_j} \epsilon A \cap E$ and hence, $x^{k_j} \epsilon A^*$

for all j large enough. Thus, the conclusion follows for this case, also.

<div align="right">Q. E. D.</div>

The equivalent problems (A) and (B) can also be formulated as

(M)   minimize $f(x)$ s. t. $x \epsilon R \cap Q$,

where $R \equiv \{x | g_i(x) \geq 0, i = 1, \ldots, m\}$ and $Q \equiv \{x | g_j(x) \geq 0, j = m+1, \ldots, q\}$. If there are any equality constraints, we assume these appear among the last $q - m$ constraints (i. e., they are involved only in defining Q and not in defining R). However, Q may contain inequality constraints as well.

It is assumed that $R^o \neq \emptyset$ and $Q \neq \emptyset$. It is desired to solve (M) by generating a sequence of unconstrained minima of an auxiliary function in such a manner that the sequence <u>must</u> be restricted to R^o, but <u>need not</u> be restricted to Q.

As throughout, we tacitly assume that $x \epsilon E^n$ and the functions of problem (M) are continuous. The generalized auxiliary function is defined as follows.

Defining Properties of $V(x, r, t)$.

  i) $V(x, r, t)$ is continuous for $x \epsilon R^o$, for any $r > 0$ and $t > 0$;

  ii) If $\{x^k\} \subset R^o$ and $x^k \to y \epsilon R - R^o$, where $\|y\| < \infty$, then

$$\lim_{k \to \infty} V(x^k, r, t) = +\infty \text{ for any } r > 0, \quad t > 0;$$

  iii) If $\{x^k\} \subset R^o$, $r_k > 0$ and $t_k > 0$ for every k, and $(x^k, r_k, t_k) \to$
   $\to (y, 0, +\infty)$, with $\|y\| < \infty$, then $\liminf_k V(x^k, r_k, t_k) \begin{cases} = +\infty \text{ if } y \notin Q, \\ \\ \geq f(y) \text{ otherwise;} \end{cases}$

iv) If $y \in R^o \cap \Omega$, $r_k > 0$ and $t_k > 0$ for every k, and $(r_k, t_k) \to (0, +\infty)$,

then $\lim_{k \to \infty} V(y, r_k, t_k) = f(y)$.

Realizations of $V(x, r, t)$ follow in the ensuing development.

With $V(x, r, t)$ so defined, a sequential technique having the desired properties can be validated under suitable conditions. Let $A \equiv \{x \mid f(x) = v* \text{ and } x \text{ is a local solution of } (M)\}$.

Theorem 2.1. Convergence of V-minima to Local Solutions of (M).

If the functions of (M) are continuous, $V(x, r, t)$ is as above, $R^o \cap \Omega \neq \emptyset$, $R \cap \Omega = \overline{R^o \cap \Omega}$, A^* is a nonempty compact isolated set of A, $r_k > 0$ and $t_k > 0$ for every k with $(r_k, t_k) \to (0, +\infty)$, then: (a) there exists a compact set S as given in Lemma 2.2, so that $S^o \supset A^*$ and, for all k large enough, the unconstrained minima $\{x^k\}$ of $\{V(x, r_k, t_k)\}$ in $R \cap S$ exist and every limit point of the uniformly bounded sequence $\{x^k\}$ is in A^*, and (b) $V(x^k, r_k, t_k) \to v*$.

Proof. Let $V(x^k, r_k, t_k) = \inf_{R \cap S} V(x, r_k, t_k)$, where S is the compact set defined in Lemma 2.2. Recall that this also means that $R \cap S^o \supset A^*$. Since $A^* \subseteq R \cap \Omega$ and $R \cap \Omega = \overline{R^o \cap \Omega}$, it follows that $R^o \cap S \neq \emptyset$.

By property (i), $V(x, r, t)$ is continuous in $R^o \cap S$ for $r > 0$ and $t > 0$. By (ii), $V(x, r, t) \to +\infty$ as $x \to y$ if $x \in R^o$ and $y \in (R - R^o)$, for any $r > 0$, $t > 0$. These facts imply that $\{x^k\} \subset R^o \cap S$. The fact that, for any $r_k > 0$, $t_k > 0$, x^k exists minimizing $V(x, r_k, t_k)$ in the compact set $R \cap S$ follows from Lemma 2.1.

Since $R \cap S$ is compact, $\{x^k\}$ must have a convergent subsequence which, for convenience, shall still be denoted by $\{x^k\}$. Hence, we can assume $x^k \to y^o \in R \cap S$. In particular, this also means

$\|y^o\| < \infty$. We shall show that $y^o \in A^*$.

By hypothesis, $R \cap Q$ is the closure of $R^o \cap Q$ and hence, by the continuity of $f(x)$, there exists $x^o \in R^o \cap Q \cap S$ such that $f(x^o) \leq v^* + \epsilon$, any $\epsilon > 0$. By definition of x^k,

$$(2.1) \qquad V(x^k, r_k, t_k) \leq V(x^o, r_k, t_k)$$

and now property (iv) implies that

$$(2.2) \qquad \liminf_k V(x^k, r_k, t_k) \leq \liminf_k V(x^o, r_k, t_k) = f(x^o) \leq v^* + \epsilon$$

for any $\epsilon > 0$, by suitable choice of x^o.

If $y^o \notin Q$ then, by property (iii), $\lim_{k \to \infty} V(x^k, r_k, t_k) = +\infty$, which contradicts (2.2). Thus, $y^o \in R \cap S \cap Q$.

If $y^o \notin A^*$ then, by definition of S, it follows that $f(y^o) \geq v^* + \lambda$, for some $\lambda > 0$. By property (iii) this means that $\liminf_k V(x^k, r_k, t_k) \geq f(y^o) \geq v^* + \lambda$, which again contradicts (2.2), since we can select x^o above such that $\epsilon < \lambda$. Consequently, we must have $y^o \in A^*$.

Further, since $\{x^k\} \subset R^o$, $x^k \to y^o$ and $y^o \in A^* \subset R \cap S^o$, this also means that for k large enough, $x^k \in R^o \cap S^o$. Thus, the x^k are uniformly bounded unconstrained minima of $V(x, r_k, t_k)$ in $R^o \cap S^o$ for large k, and every limit point of $\{x^k\}$ is in A^*.

For part (b), since $y^o \in A^*$, (2.2) and property (iii) imply that

$$(2.3) \qquad v^* = f(y^o) \leq \liminf_k V(x^k, r_k, t_k) \leq v^* + \epsilon,$$

for any $\epsilon > 0$. Similarly, (2.1) and properties (iii) and (iv) yield

(2.4) $v^* \leq \lim \sup_{k \to \infty} V(x^k, r_k, t_k) \leq v^* + \epsilon$,

for any $\epsilon > 0$. From (2.3) and (2.4), $\lim_{k \to \infty} V(x^k, r_k, t_k) = v^*$.

<div align="right">Q. E. D.</div>

The following corollary gives the global result indicated in the previous discussion. It will be apparent that this result follows immediately from the theorem and the fact that the set of global minima of (M) must be closed under our assumptions on the problem functions. The second corollary is an obvious consequence.

Corollary 1. Convergence to Bounded Set of Global Minima.

Assuming the hypotheses of Theorem 2.1, if the set A^* of global minima of problem (M) is bounded, then the conclusions of the theorem hold and the sequence of unconstrained local minima $\{x^k\}$ of $\{V(x, r_k, t_k)\}$ in $R \cap S$ is such that every limit point is a global minimum of (M).

Corollary 2. Convergence to Isolated Local Minimum.

Assuming the hypotheses of Theorem 2.1, if $A^* = \{x^*\}$ and x^* is an isolated local minimum of (M), then the conclusions of the theorem hold and, furthermore, the sequence of unconstrained local minima $\{x^k\}$ of $\{V(x, r_k, t_k)\}$ in $R \cap S$ is such that $\{x^k\}$ itself converges to x^*.

Before proceeding with the general development, a few interesting facts may be noted about the general auxiliary function that has been defined. Suppose this function has the form $V[f(x), g_1(x), \ldots, g_q(x), r, t]$, V is once differentiable in $f, g_1, \ldots, g_q$, and these latter functions are differentiable in x, for $x \in R^o$ and any $r > 0, t > 0$. Then, if x^k is a local unconstrained minimum of V, denoting the corresponding value of V

by V^k, it follows that

$$\nabla V^k = \frac{\partial V}{\partial f(x^k)} \nabla f(x^k) + \sum_{i=1}^{q} \frac{\partial V}{\partial g_i(x^k)} \nabla g_i(x^k) = 0 .$$

This equation is the same as $\nabla L(x^k, u^k) = 0$, where L is the Lagrangian associated with Problem (M), providing we assume $\partial V/\partial f(x^k) \neq 0$ and we set

$$u^k = \frac{-\partial V/\partial g_i(x^k)}{\partial V/\partial f(x^k)} .$$

This shows how we can be led to establishing a direct connection between the conditions that hold at a local minimum of the penalty function and the conditions that hold at a local solution of (M).

The above also indicates the connection with duality in the convex programming problem that will be briefly summarized in Section 5. If V is convex increasing in the convex function f and decreasing in the concave functions $g_i(x)$, $i = 1, \ldots, q$, then it follows that $V(x, r, t)$ is a convex function in R^o. If we further assume $\partial V/\partial f > 0$, then it follows immediately that (x^k, u^k) is a dual feasible point.

Thus, the connections between the conditions associated with minimization of the penalty function and the optimality and duality conditions are apparent simply from the structure of this function, our defining properties having assumed very little concerning its particular form.

3. Generalized Interior and Exterior Auxiliary Functions

Returning to the general development, we wish to be able to solve Problem (M) when we require only that $x \in R$, or that $x \in Q$, i.e., we wish to solve the problems

M(R)   minimize $f(x)$ subject to $x \in R$,

and

M(Q)   minimize $f(x)$ subject to $x \in Q$,

using an auxiliary function technique analagous to that given above for Problem (M).

The Problems M(R) and M(Q) have the same basic structure, the essential difference being that we shall assume $R^o \neq \phi$ and $Q \neq \phi$ (so that we may have $Q^o = \phi$). The remaining distinction is procedural: we insist on restricting x to R^o in the course of solving M(R), whereas x need not be restricted to Q in the course of solving M(Q).

Note that M is the same as M(R) if $Q = E^n$ and M is the same as M(Q) if $R = E^n$. Since there were no restrictions on R and Q in the above development (other than $R^o \neq \phi$, $Q \neq \phi$), the above convergence theorem for Problem (M) utilizing $V(x, r, t)$ is valid for $R = E^n$ or $Q = E^n$. We shall modify the respective definitions of $V(x, r, t)$ in that we shall associate the parameter r with Problem M(R) and the parameter t with Problem M(Q).

We are thus led to the auxiliary functions for M(R) and M(Q) which we shall call $U(x, r)$ and $T(x, t)$, respectively.

We assume $R^o \neq \phi$ and essentially arrive at the defining properties of $U(x, r)$ by setting $Q = E^n$ and by suppressing t in the definition of $V(x, r, t)$.

Defining Properties of $U(x, r)$.

    a) $U(x, r)$ is continuous for $x \in R^{o}$, for any $r > 0$;

    b) If $\{x^{k}\} \subset R^{o}$ and $x^{k} \to y \in R - R^{o}$, where $\|y\| < \infty$, then
$\lim_{k \to \infty} U(x^{k}, r) = +\infty$ for any $r > 0$;

    c) If $\{x^{k}\} \subset R^{o}$, $r_{k} > 0$ for every k, and $(x^{k}, r_{k}) \to (y, 0)$, where
$\|y\| < \infty$, then $\lim \inf_{k} U(x^{k}, r_{k}) \geq f(y)$;

    d) If $y \in R^{o}$, $r_{k} > 0$ for every k, and $r_{k} \to 0$, then
$\lim_{k \to \infty} U(y, r_{k}) = f(y)$.

Assuming $Q \neq \emptyset$, setting $R = E^{n}$ and suppressing r in the
definition of $V(x, r, t)$, we obtain the defining properties of $T(x, t)$.

Defining Properties of $T(x, t)$.

    A) $T(x, t)$ is continuous for $x \in E^{n}$, for any $t > 0$;

    B) If $\{x^{k}\} \subset E^{n}$, $t_{k} > 0$ for every k, and $(x^{k}, t_{k}) \to (y, +\infty)$, where
$\|y\| < \infty$, then $\lim \inf_{k} T(x^{k}, t_{k}) \begin{cases} = +\infty \text{ if } y \notin Q, \\ \geq f(y) \text{ otherwise;} \end{cases}$

    C) If $y \in Q$, $t_{k} > 0$ for every k, and $t_{k} \to +\infty$, then
$\lim_{k \to \infty} T(y, t_{k}) = f(y)$.

As direct consequences of these definitions and the previous results,
we obtain methods for solving $M(R)$ and $M(Q)$.

Theorem 3.1. Convergence of U-Minima to Local Solutions of M(R).

    The convergence theorem for $V(x, r, t)$ applied to (M) is valid if

    $V(x, r, t)$ is replaced by $U(x, r)$, (M) is replaced by $M(R)$, $Q = E^{n}$,

    and t is suppressed, in the statement of that theorem.

Theorem 3.2. Convergence of T-Minima to Local Solutions of M(Q).

    The convergence theorem for $V(x, r, t)$ applied to (M) is valid

    if $V(x, r, t)$ is replaced by $T(x, t)$, (M) is replaced by $M(Q)$,

    $R = E^{n}$, and r is suppressed, in the statement of that theorem.

The steps in the proofs of these two theorems are precisely the same as those in the V-function theorem, with obvious modifications, on making the indicated substitutions.

4. Additional Results for General Interior and Exterior Penalty Functions

We obtain a realization of $V(x, r, t)$ in terms of $U(x, r)$ and $T(x, t)$ by defining $V_1(x, r, t) \equiv U(x, r) + T(x, t) - f(x)$. As required by the definition of $V(x, r, t)$, we assume the problem functions continuous, $R^O \neq \emptyset$, and $\Omega \neq \emptyset$. The following result is a direct consequence of the defining properties.

Lemma 4. 1. $V_1(x, r, t)$ is a V-function (i. e. , a function $V(x, r, t)$ satisfying
(i) through (iv) above.)

"Penalty" functions used in this approach are usually obtained by adding to $f(x)$ a "penalty term" which absorbs the effects of the constraints of the given problem and the involved parameter. Towards developing these functions as particular realizations of the functions $U(x, r)$ and $T(x, t)$ defined above, we define the following. Assume $R^O \neq \emptyset$ and $\Omega \neq \emptyset$.

Defining Properties of $I(x, r)$.

a_1) $I(x, r)$ is continuous for $x \in R^O$, for any $r > 0$;

b_1) If $\{x^k\} \subset R^O$ and $x^k \to y \in R - R^O$, where $\|y\| < \infty$, then
$\lim_{k \to \infty} I(x^k, r) = +\infty$ for any $r > 0$;

c_1) If $\{x^k\} \subset R^O$, $r_k > 0$ for every k, and $(x^k, r_k) \to (y, 0)$, where $\|y\| < \infty$, then $\liminf_k I(x^k, r_k) \geq 0$;

d_1) If $y \in R^O$, $r_k > 0$ for every k, and $r_k \to 0$, then
$\lim_{k \to \infty} I(y, r_k) = 0$.

Defining Properties of $0(x, t)$.

A_1) $0(x, t)$ continuous for $x \in E^n$, for any $t > 0$;

E_1) If $\{x^k\} \subset E^n$, $t_k > 0$ for every k, and $(x^k, t_k) \to (y, +\infty)$, where

$\|y\| < \infty$, then $\liminf_k 0(x^k, t_k)$ $\begin{cases} = +\infty \text{ if } y \notin \Omega \\ \geq 0 \text{ otherwise;} \end{cases}$

C_1) If $y \in \Omega$, $t_k > 0$ for every k, and $t_k \to +\infty$, then

$\lim_{k \to \infty} 0(y, t_k) = 0$.

With "U-function" and "T-function" meaning $U(x, r)$ and $T(x, t)$, respectively, as defined in the previous section, we obtain the following direct consequences.

Lemma 4.2. $U_1(x, r) \equiv f(x) + I(x, r)$ is a U-function.

Lemma r. 3. $T_1(x, t) \equiv f(x) + 0(x, t)$ is a T-function.

The proofs of the lemmas follow immediately from the continuity of $f(x)$ and the defining properties.

A further immediate consequence of the three lemmas above is the following.

Corollary 1. $V_2(x, r, t) \equiv U_1(x, r) + T_1(x, t) - f(x) = f(x) + I(x, r) + 0(x, t)$

is a V-function.

This provides a realization of the V-function, associated with Problem (M), in terms of the objective function $f(x)$ of that problem, and the penalty functions associated with problems $M(R)$ and $M(\Omega)$ (through $U_1(x, r)$ and $T_1(x, t)$ above).

The penalty functions defined in [5] for the problems $M(R)$, $M(\Omega)$, and (M) are, respectively,

$$U_2(x, r) \equiv f(x) + s(r) I(x),$$

$$T_2(x, t) \equiv f(x) + p(t) 0(x),$$

and $\quad V_3(x, r, t) \equiv f(x) + s(r) I(x) + p(t) 0(x).$

The penalty functions utilized to date generally subscribe to the above forms. We shall summarize the defining properties of

$s(r)$, $I(x)$, $p(t)$, and $0(x)$. It follows readily that $s(r)$ $I(x)$ is an I-function (satisfies properties (a_1) through (d_1) above), and $p(t)$ $0(x)$ is an 0-function (satisfies (A_1) through (C_1) above).

Defining Properties of $I(x)$ and $s(r)$ [5].

1) $I(x)$ is continuous for $x \in R^0$;

2) If $\{x^k\} \subset R^0$ and $x^k \to y \in R - R^0$, then $\lim_{k \to \infty} I(x^k) = +\infty$;

3) $s(r)$ is a (scalar-valued) function of r, continuous for $r > 0$;

4) If $r_1 > r_2 > 0$, then $s(r_1) > s(r_2) > 0$;

5) If $r_k > 0$ for every k and $r_k \to 0$, then $\lim_{r_k \to \infty} s(r_k) = 0$.

Defining Properties of $0(x)$ and $p(t)$ [5].

1^1) $0(x)$ continuous for $x \in E^n$;

2^1) $0(x)$ $\begin{cases} = 0 \text{ if } x \in C, \\ > 0 \text{ if } x \notin C; \end{cases}$

3^1 $p(t)$ is a (scalar-valued) function of t, continuous for $t > 0$;

4^1) If $t_2 > t_1 > 0$, then $p(t_2) > p(t_1) > 0$;

5^1) If $t_k > 0$ for every k and $t_k \to +\infty$, then $\lim_{k \to \infty} p(t_k) = +\infty$.

As straightforward consequences of these defining properties we obtain the following:

Lemma 4. 4. $I_1(x, r) \equiv s(r) I(x)$ is an I-function.

Lemma 4. 5. $0_1(x, t) \equiv p(t) 0(x)$ is an 0-function.

In view of the above results, this means that U_2 is a U-function (i. e., satisfies (a) through (d) above), T_2 is a T-function (i. e., satisfies (A) through (C) above) and V_3 is a V-function (i. e., satisfies (i) through (iv) above.)

From this point on, we shall be dealing only with the U_2, T_2, and V_3 functions defined above. For convenience, therefore, we shall revert to the notation, U, T, and V for these functions.

It is first noted from the above development that these penalty functions are special cases of the general auxiliary functions defined in Sections 2 and 3, so that the respective convergence theorems apply. Additional characterizations of convergence follow as corollaries of these theorems, from the particular structure of these functions.

In the following we assume the conditions of Theorems 3.1 and 3.2 are satisfied.

We know $\{x^k\} \subset R^o \cap S^o$ for k large, where

$$\min_{R \cap S} U(x, r_k) = U(x^k, r_k) ,$$

and S is the compact set defined in Lemma 2.2. Recall that $S^o \supset A^*$, a compact set of local minima associated with the value $f(x) = v^*$. Also, we know there exists at least one limit point x^* of $\{x^k\}$, and any such limit point is in A^*. The following results show that for $U(x, r)$ as presently defined, we can assure that the minimizing points yield values of f, U and I that converge monotonically.

Corollary 1 of Theorem 3.1.

If $U(x, r) = f(x) + s(r) I(x)$, where $I(x)$ and $s(r)$ satisfy properties (1) - (5) given above, $f(x)$ is continuous, $\{r_k\}$ is a positive strictly decreasing null sequence, and $\min_{R \cap S} U(x, r_k) = U(x^k, r_k)$, then $f(x^k) \downarrow v^*$, $I(x^k) \uparrow \nu$, where $\nu = +\infty$ if $x^* \in (R-R^o)$ and $|\nu| < \infty$ if $x^* \in R^o$, and $s(r_k) I(x^k) \to 0$. Also, if $I(x) \geq 0$ in $R^o \cap S$, then $U(x^k, r_k) \downarrow v^*$.

Proof. By definition of x^k, $U(x^k, r_k) \leq U(x, r_k)$ for every $x \in S \cap R$, so that

(4.1) $\qquad f(x^k) + s(r_k) I(x^k) \leq f(x^{k+1}) + s(r_k) I(x^{k+1})$, and

(4.2) $\qquad f(x^{k+1}) + s(r_{k+1}) I(x^{k+1}) \leq f(x^k) + s(r_{k+1}) I(x^k)$.

Multiplying the first inequality by $s(r_{k+1})$, the second by $s(r_k)$,

summing and rearranging yields

$$[s(r_k) - s(r_{k+1})] f(x^{k+1}) \leq [s(r^k) - s(r^{k+1})] f(x^k).$$

Since $r_k > 0$ and $r_k \downarrow 0$ strictly, $s(r_k) > s(r_{k+1})$ by definition of $s(r)$, so that $f(x^{k+1}) \leq f(x^k)$. This, together with (4.1), also implies that $I(x^{k+1}) \geq I(x^k)$.

Since we also have $f(x) \geq v^*$ for $x \in S \cap R$, we must have $f(x^k) \downarrow \bar{v} \geq v^*$. We know that at least one limit point of $\{x^k\}$ exists, and Theorem 3.2 assures us that limit points of $\{x^k\}$ must be in A^*, so that we must have $\bar{v} = v^*$; i.e., $f(x^k) \downarrow v^*$.

The conclusions regarding the convergence of $I(x^k)$ follow from the monotonicity shown above and from the properties (1) and (2) of $I(x)$.

By property (c) of the U function and by the result for $U(x, r)$ analagous to equation (2.1), it follows that

$$v^* \leq \lim_{j \to \infty} U(x^{k_j}, r_{k_j}) \leq v^* + \epsilon$$

for <u>any</u> $\epsilon > 0$ and <u>any</u> convergent subsequence $\{x^{k_j}\}$ of $\{x^k\}$. Hence, $U(x^{kj}, r_{k_j}) \to v^*$. It follows that $U(x^k, r_k) \to v^*$, by the compactness of $R \cap S$ and the fact that at least one convergent subsequence exists.

From the above results, we obtain immediately that $s(r_k) I(x^k) \to 0$.

Finally, if $I(x) \geq 0$ in $R^0 \cap S^0$, then it follows that $U(x, r_{k+1}) \leq U(x, r_k)$ for $r_k > r_{k+1} > 0$ and any $x \in R \cap S$. This implies $U(x^{k+1}, r_{k+1}) \leq U(x^k, r_k)$ and using the result shown above, $U(x^k, r_k) \downarrow v^*$.

Q. E. D.

The analagous results for the $T(x, t)$ function can be proved in a similar manner, and are summarized as follows.

Corollary 1 of Theorem 3.2.

If $T(x, t) = f(x) + p(t) O(x)$, where $O(x)$ and $p(t)$ satisfy properties $(1^1) - (5^1)$ given above, $f(x)$ is continuous, $\{t_k\}$ is a positive strictly increasing unbounded sequence, and $\min_S T(x, t_k) = T(x^k, t_k)$, then $T(x^k, t_k) \Uparrow v*$, $f(x^k) \uparrow v*$, $O(x^k) \downarrow 0$, and $p(t_k) O(x^k) \to 0$.

5. Global and Dual Results for Convex Programming[1]

For the following development, we utilize the formulation

(C)   minimize $f(x)$ subject to $g_i(x) \geq 0$, $i = 1, \ldots, m$.

We shall summarize in this section some of the principal results that hold when the interior and exterior penalty function methods are applied to solve (C), when this is a convex programming problem. These methods are specialized to take advantage of the convexity assumption, essentially by defining the penalty functions in such a way that these are themselves convex functions. The theorems given in the previous section are of course still valid, essentially with "local" replaced by "global." Furthermore, duality results are forthcoming that provide additional information.

We first discuss the interior point method. It is assumed that $R^o = \{x \,|\, g_i(x) > 0, \; i = 1, \ldots, m\} \neq \emptyset$. We shall further assume the penalty factor $I \equiv I[g(x)]$, where $g \equiv (g_1, \ldots, g_m)$, and that $h(r) = r$, so that the form of the interior penalty function is now $U(x, r) \equiv f(x) + rI[g(x)]$. Recall that $I[g(x)]$ is continuous in R^o and, with $\{x^k\} \subset R^o$ and $x^k \to y \in (R - R^o)$, $I[g(x^k)] \to +\infty$.

It is possible to assure the convexity of $U(x, r)$ in R^o by appropriately defining $I[g(x)]$. The following lemma is immediate and yields a suitable definition.

Lemma 5.1. Convexity of $I[g(x)]$.

If $I(g)$ is a convex decreasing function of g for $g > 0$, and the $g_i(x)$ are concave, then $I[g(x)]$ is a convex function of x in R^o.

---

[1]The material in this section is essentially a summary of some of the results obtained in [5] and is presented here for completeness. Most importantly, this development shows how the penalty function can be used to take advantage of the convexity assumption to yield global and dual results.

It is readily verified, for example, that the functions $-\Sigma \ell n g_i$ and $\Sigma 1/g_i$ satisfy the conditions of the lemma. Assuming the hypothesis of the lemma holds, then $I\lceil g(x) \rceil$ is convex, and hence, $U(x, r) = f(x) + r I[g(x)]$ is convex in R^o, for $r > 0$.

The following theorem then holds for $U(x, r)$ as defined. Note that the conclusions are immediately implied by the interior-point auxiliary function Theorem 3.1, and Corollary 1 (Section 4) of that theorem, once the convexity assumptions are introduced.

Theorem 5.1. Interior Penalty Method Convergence to Solution of Convex

Programming Problem [5].

If the solutions of the convex programming problem (C) are bounded, $R^o = \{x \mid g_i(x) > 0, i = 1, \ldots, m\} \neq 0$, I(g) is defined, decreasing and convex in g when $g > 0$ and $\{r_k\} \downarrow 0$, then $U(x, r)$ is a convex function in R^o for $r > 0$, for r_k small there exists $x(r_k)$ that minimizes $U(x, r_k)$ in R^o, every local minimum of $U(x, r_k)$ in R^o is a global minimum in R^o, any limit point $x*$ of $\{x(r_k)\}$ solves (C), $U\lceil x(r_k), r_k \rceil \downarrow v*$, $\{f\lceil x(r_k) \rceil\} \downarrow v* \equiv \min_R f(x)$, and $I\lceil x(r_k) \rceil \uparrow \gamma$, where $\gamma = +\infty$ if $x* \in (R - R^o)$, and $|\gamma| < \infty$ if $x* \in R^o$, and $r_k I\lceil x(r_k) \rceil \to 0$.

As expected from the previous general theorem, Theorem 3.1, and the convexity assumption, the minimizing points are now global and all limit points are global solutions of (C). The monotone convergence results are also global. These results follow from the convexity of $U(x, r)$ and the well known fact that any local solution of the convex programming problem (C) is also a global solution. Hence, the compact set $A*$ of local minima of (C) is the unique set of global minima, and the compact set S, utilized

for our previous local results, can be taken to be <u>any</u> compact set S such that $S^o \supset A^*$.

In effect, the role of convexity is essentially to make the previous theorem a "global" one. It should be mentioned that in [5] a condition is given that regularizes $I[g(x)]$ such that, for the convex problem the minimizing $x(r_k)$ exists for <u>all</u> $r_k > 0$, not merely for r_k small. A similar condition has not been obtained for the general problem.

A "dual" of problem (C) was formulated by Wolfe [11], assuming the problem functions are once differentiable. The Lagrangian function associated with problem (C) is defined as $L(x, u) \equiv f(x) - \sum_{i=1}^{m} u_i g_i(x)$. The dual of (C) is then defined as,

(D)   maximize $L(x, u)$

subject to $\nabla_x L(x, u) = 0$, $u_i \geq 0$, $i = 1, \ldots, m$.

The dual relationship is defined in the sense given in the following two theorems.

Theorem 5.2. Primal-Dual Bounds.

If (C) is a convex programming problem, y is any primal feasible point, and (x, u) is any feasible point of (D), then $f(y) \geq L(x, u)$.

Theorem 5.3. Equality of Primal-Dual Values at Optimal Solutions.

If (C) is a convex programming problem and the Kuhn-Tucker constraint qualification (see Ref. [6]) holds at a solution x* of (C), then there exists a solution (x^*, u^*) of (D) and $f(x^*) = L(x^*, u^*)$.

The nondifferentiable form of (D) is given by assuming only continuity and replacing $\nabla_x L(x, u) = 0$ by $L(x, u) = \inf_\xi L(\xi, u)$. The first dual theorem goes through easily with this change. By making this substitution and

assuming $R^o \equiv \{x \,|\, g_i(x) > 0, \quad i = 1, \ldots, m\} \neq \emptyset$, it is possible to dispense

with the differentiability requirement and the Kuhn-Tucker constraint

qualification, and arrive at a non-differentiable version of the second dual

theorem given above, assuming only continuity.

In conjunction with the Kuhn-Tucker constraint qualification, it is

relevant to point out the fact that this qualification is satisfied providing

the constraints are differentiable and convex, and the interior of the

constraint region defined by the nonlinear constraints is nonempty [13].

We now turn briefly to the significant duality results that follow from

the $U(x, r)$ method. For this development, we give U additional structure,

and now assume that the penalty factor has the form $\sum\limits_{i=1}^{m} I_i[g_i(x)]$. The

result applies to the nondifferentiable form of the dual problem,

$(\overline{D})$ maximize $L(x, u)$ subject to $L(x, u) = \inf_\xi L(\xi, u)$, $u_i \geq 0$, $i = 1, \ldots, m$.

Theorem 5.4. Dual-Feasibility and Convergence for Interior-Point Methods [5].

Assume the conditions of Theorem 5.1. If, in addition, each $I_i(g_i)$

is differentiable in g_i when $g_i > 0$, then for r_k small enough

$[x(r_k), u(r_k)]$ satisfies the constraints of $(\overline{D})$ where

$u_i(r_k) \equiv -r_k \dfrac{\partial I_i \{g_i [x(r_k)]\}}{\partial g_i}$, $i = 1, \ldots, m$. Furthermore,

$\lim_{k \to \infty} L[x(r_k) , u(r_k)] = v^* = \min_R f(x)$ so that all limit points of

$\{\lceil x(r_k), u(r_k)\rceil\}$ solve $(\overline{D})$.

The definition of the quantities $u_i(r_k)$ and the result given above

is most readily appreciated by assuming the functions differentiable,

differentiating $U(x, r)$ directly, and comparing the result to $\nabla_x L(x, u)$, which

would appear in the differentiable form of $(\overline{D})$.

The above result is a significant one, not generally shared by other methods. One very important application of it is that upper and lower bounds on the optimal value v* are generated with the determination of each minimizing $x(r_k)$. This is a direct consequence of the first dual theorem given above. It provides a natural convergence criterion. Also, the Lagrange multipliers are critically involved in providing optimality conditions. In effect, the dual theorem provides an explicit relationship between $U(x, r)$ and the Lagrangian $L(x, u)$ associated with problem (C).

A parallel development for the exterior-point method for the convex problem is given in [5]. Again, similar conditions are invoked on the form of $T(x, t)$, assuring its convexity. The convergence theorem is essentially a "global" version of the general exterior-point theorem given in Section 3. Analagous to the development for the $U(x, r)$ interior method, a condition is given in [5] that guarantees the existence of the minimum of $T(x, t)$ for all $t > 0$, for the convex problem. Finally, a comparable dual theorem is proved. In view of the fact that the optimal value is approached from below in this method, the dual result provides only a lower bound on the optimal value.

Recalling that $R^o \neq \emptyset$ is not required for the feasible-exterior $T(x, t)$ - function method, the above indicated results imply, in particular, that the Kuhn-Tucker constraint qualification need not be satisfied for problem (C). This means, when conditions assure that a minimizing sequence of $T(x, t_k)$ leads to a solution of the problem, then we have a way of characterizing such a solution although the constraint qualification may not hold.

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Our objective is to provide a more general theoretical basis for those
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based on successive unconstrained minimizations of a parametric
auxiliary function. In order to do this, we first give basic defining
properties of a general auxiliary function and obtain a proof of local
convergence for the mildly regulated nonconvex problem. We proceed
from this general result to develop a family of auxiliary functions by
giving the general function more and more structure, eventually being
led to the general form of the "penalty" functions that conventionally
have been utilized. Similarly, the problem structure is increasingly
specialized and we deal finally with the convex problem. As expected,
stronger results are obtained as additional structure is assumed.

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